



ISSN : 2347-2251

**Indo-American Journal of
Pharma and Bio Sciences**



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Recent Trends in Computational Intelligence: Paradigms and Applications

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Abstract: To be intelligent is to be able to think and reason, to be able to grasp, understand, and benefit from one's own experience. An umbrella term for the three basic technologies of artificial neural networks, fuzzy systems and evolutionary computing is computational intelligence (CI). These clever algorithms are a component of the area of artificial intelligence, along with logic, deductive reasoning, expert systems, case-based reasoning, and symbolic machine learning systems (AI). For example, computer science, philosophy, sociology, and biology are all examples of study fields that may be combined in this way. It is the study of adaptive processes that permit or assist intelligent behaviour in complex and changing contexts (CI). These are AI paradigms that are able to learn or adapt to new contexts, to generalise, abstract, discover, and associate with one other. The CI paradigms of artificial neural networks, evolutionary computation, swarm intelligence, artificial immune systems, and fuzzy systems are among the CI paradigms to be explored further. Since no one paradigm is better to the others in all scenarios, the current tendency is to construct hybrids of various Computational Intelligence (CI) approaches, which have proven useful in solving real-world issues. Our goal is to take use of each component's strengths while also eliminating its flaws in the hybrid Computational Intelligence (CI) system. Computational Intelligence and its concepts and applications are discussed in this article.

Keywords: Artificial neural networks, fuzzy systems, evolutionary computation, and symbolic machine learning systems all fall under the umbrella term "computational intelligence."

INTRODUCTION

The building of algorithmic models to tackle more complicated problems is a fundamental drive in algorithmic development. The modelling of biological and natural intelligence has yielded enormous advances, leading in so-called "intelligent systems." Artificial neural networks, evolutionary computation, swarm intelligence, artificial immune systems, and fuzzy systems are some of the more intelligent algorithms on this list. Symbolic machine learning and expert systems, as well as logic and deductive reasoning, also fall under the umbrella of artificial intelligence (AI).

PARADIGMS OF COMPUTER INTELLIGENCE:

Computational Intelligence (CI) is characterised by the following five basic paradigms:

Neural networks based on artificially intelligent systems (NN)

Evolving computations (EC)

The intellect of the horde (SI)

Computerized immune systems (AIS)

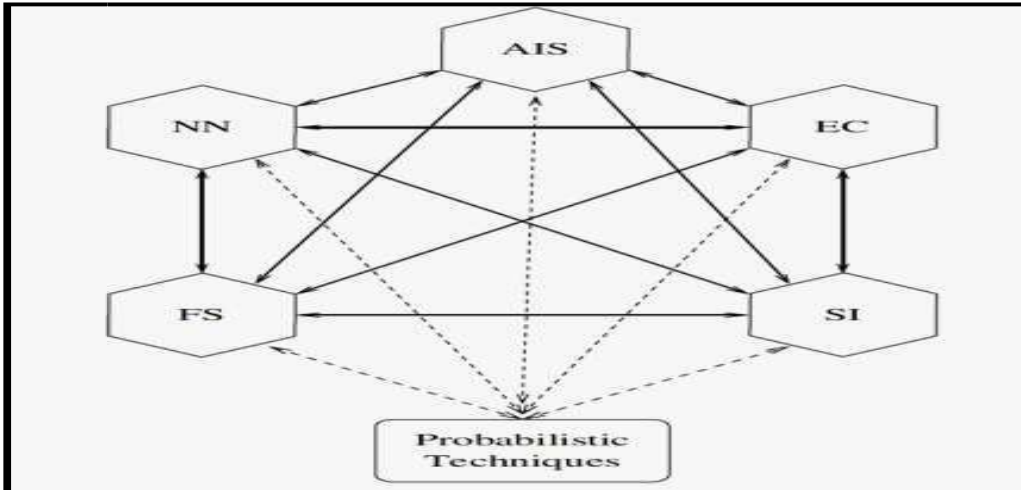
Uncertain systems (FS).

Probabilistic approaches are often employed in conjunction with Computational Intelligence as seen in Figure 1.

(CI) procedures Lotfi Zadeh invented the phrase "soft computing" to describe an amalgamation of the various

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Computational Intelligence (CI) paradigms and probabilistic approaches. Techniques from many paradigms may be merged to create hybrid systems, as the arrows suggest.

Fig.1. Computational Intelligence Paradigms

In each of the CI paradigms, there is a biological basis for its development. Neural networks (NNs), evolutionary computation (EC), social networks (SI), the human immune system (AIS), and interactions with the environment (FS) are some of the models used to understand how live creatures interact with their environments. Machine Learning and the Use of Artificial Neural Networks

The brain is a nonlinear, parallel computer that is complicated and nonlinear. When compared to computers, it is able to carry out activities such as pattern recognition, vision, and motor control at a far quicker rate than silicon gates and neural systems. Research into algorithmic modelling of biological brain systems, referred to as artificial neural networks, was inspired by the capacity to learn, remember and yet generalise (NN).

The human brain is thought to have between 10 and 500 billion neurons, with 60 trillion connected to each other.

synapses. The neurons are organised in a certain way.

It is estimated that there are around 1000 main modules.

consisting of around 500 nerve cells

networks. Small artificial NNs with a narrow focus on a single problem are now seeing success in neural modelling.

NNs of a reasonable size may quickly solve problems with a single aim, as long as the current computing and storage capacity capabilities. The brain's dispersed components can address many issues at the same time. Biological neural systems are made up of nerve cells, known as neurons. Neurons are made up of a cell body, dendrites, and an axon, as seen in Fig.2. Neurons are intertwined in such a vast way that it is impossible to separate them from one other. between the axon and dendrite of one neuron and the other. This is what I mean by saying that

It is called a synapse. Dendrites transmit impulses to the axon, which then transmits signals to all associated dendrites. Only when a neuron fires does a signal go from the cell to the axon. A signal may be inhibited or excited by a neuron. It is a model of a biological neuron, an artificial neuron (AN) (BN). A signal is sent out to all linked ANs when one AN is activated, which receives signals from the environment or other ANs.

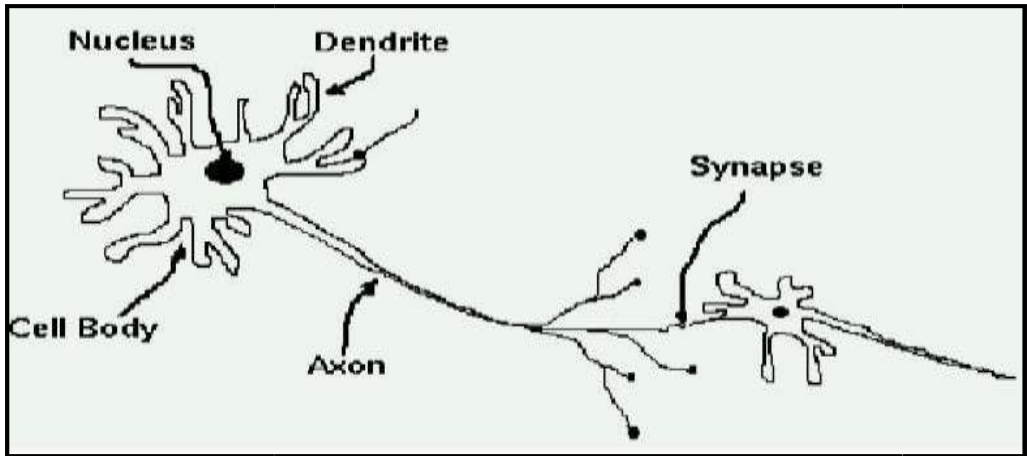
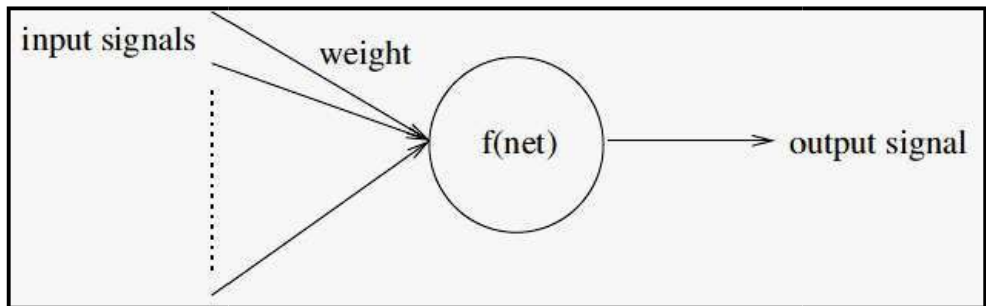


Fig.2.A Biological Neuron

An artificial neuron is shown in the following figure 3. Through the use of negative and positive numerical weights associated with each connection to the artificial neuron, input signals may be blocked or activated



(AN). To regulate the firing of an artificial neuron (AN) and the intensity of its output signal, a function is used. Artificial neurons (AN) gather all incoming signals and calculate a net input signal, which is determined by the relevant weights. To compute the artificial neuron's output signal, an activation function takes into account the net input signal (AN).

Fig.3.An Artificial Neuron

An artificial neural network (NN) is a layered network of artificial neurons. An NN may consist of an input layer, hidden layers and an output layer. Artificial neurons in one layer are connected, fully or partially, to the ANs in the next layer. Feedback connections to previous layers are also possible. A typical NN structure is depicted in the following fig.4.

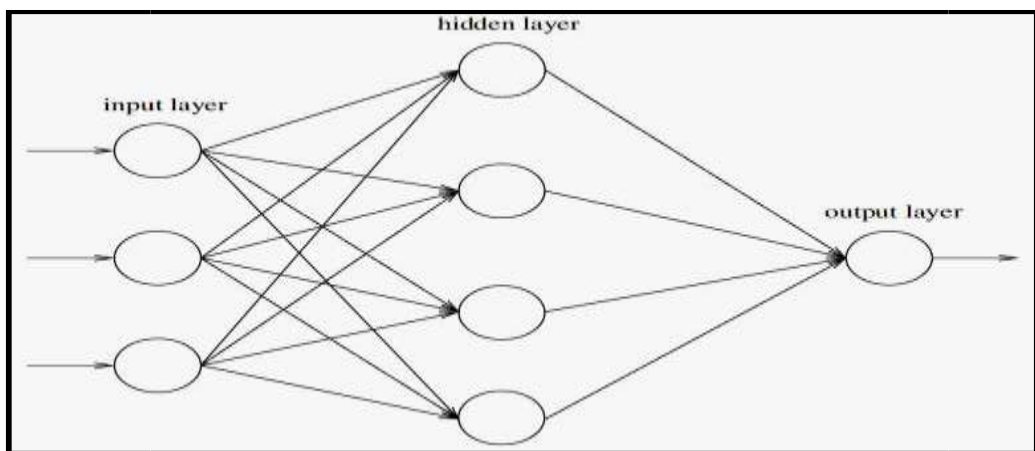


Fig.4An Artificial Neural Network

Multilayer feedforward NNs like conventional backpropagation, functional link, and product unit networks are examples of multilayer feedforward NNs, in addition to single-layer NNs like the Hopfield network.

Temporal neural networks, such as Elman and Jordan simple recurrent and time-delay neural networks.

Kohonen self-organizing feature maps and learning vector quantizers are two examples of self-organizing NNs.

Some radial basis function networks, for example, combine supervised and unsupervised NNs.

There are several uses for these NN kinds, including diagnosis, voice recognition, data mining, music composition, image processing, forecasting, credit card approval, pattern recognition, planning game strategy and compression, to name a few.

Evolving computation (EC):

This aims to emulate nature evolution, where the most important principle is survival of the fittest: weaker organisms must perish.

Individuals in an evolutionary population are referred to as "chromosomes" in evolutionary algorithms. Chromosomes are genetic markers that help identify people within a group. A gene is the name given to each trait. Genes have a number of different values, each of which is known as an allele. Individuals strive to have children in each generation. Individuals with the highest ability to survive have the best chance of reproducing. When two organisms meet, the result is a new creature, which is called a "crossover". Alleles on the chromosome may be altered through mutation in each person in the population. An individual's ability to survive is evaluated using a fitness function that represents the problem's goals and limitations.

Evolutionary Algorithms (EA) fall under the following categories:

Evolutionary algorithms

Genetic algorithm

Adaptive software design

Evolutionary approaches

[Different evolutionary processes]

Cultural change

Evolution

For example, data mining, combinatorial optimization, fault detection, classification, clustering, scheduling, and time series approximation have all been effectively implemented using evolutionary computing.

SWARM INTERACTIVE INTELLIGENCE.

This has its roots in the study of social creatures such as colonies or swarms.

As a result of research into the social behaviour (individuals) of groups of organisms (swarms), algorithms for effective optimization and clustering were developed. Particle swarm optimization and ant colony optimization techniques, for example, were developed from simulated studies of bird groups' elegant, but unpredictable, choreography.

IMMUNE SYSTEMS, NATURAL:

Non-self or antigen cells that enter the body are distinguished from those that belong to the body by this extraordinary pattern matching capacity (referred to as self). To better respond to antigens, the NIS adapts by remembering the antigen's structure and storing it in memory for future use.

Four models of the immune system can be found in NIS research: The conventional perspective of the immune system is that the immune system differentiates between self and non-self, employing lymphocytes generated in the lymphoid organs. B-Cells "learn" to recognise and bind to antigens in this manner. The clones that are created are likewise tainted.

Immune systems can tell the difference between harmful and non-threatening antigens, according to the danger hypothesis. B-Cells are supposed to form a network in network theory. When a B-cell reacts to an antigen, the network of B-cells to which it is attached is activated and stimulated. Some characteristics of a natural immune system (NIS) are modelled by an artificial immune system (AIS), which is generally used to tackle pattern recognition issues and to conduct classification jobs. Anomaly detection, including fraud and computer virus detection, is a major use of AISs.

INSANELY COMPLEX SYSTEMS

Either an element is a member of a set or it isn't, according to traditional set theory. Approximate reasoning is made possible by fuzzy sets and fuzzy logic. There is a certain level of assurance that an element is in a fuzzy collection. Assuming these facts are uncertain, fuzzy logic uses them to infer additional facts, each with a particular degree of confidence associated with them. Fuzzy sets and logic may be used to simulate common sense in certain ways. There are several applications for fuzzy systems, including gear transmission and brake systems in automobiles, elevators, household appliances, traffic signal management, and more.

CONCLUSION

Problems for which there are no efficient computing techniques are studied in Computational Intelligence (CI). These kinds of challenges arise every day in the lives of living beings, such as extracting meaning from perceptions, interpreting language, solving ill-defined computational vision problems, and surviving in a harsh environment owing to the development of the brain. For example, artificial neural networks have been utilised in a broad variety of applications ranging from illness diagnosis to voice recognition to data mining to music composition to image processing to forecasting to robot control to credit approval. Real-world uses of evolutionary computing have proved successful. An artificial immune system is mostly used for pattern detection, classification, and data clustering. Anti-fraud and computer virus detection are two of the most common uses of anomaly detection in AISs. The ultimate hurdle for computational intelligence may be cognitive robots.

REFERENCES

1. The Second Edition of Computational Intelligence: An Introduction At the end of 2007, Engelbrecht Publisher: John Wiley & Sons, Inc.
2. A. Konar, Computational Intelligence: Principles, Techniques, and Applications. Springer 2005.
3. W. Duch, Computational Intelligence: Towards a Comprehensive Foundation. "Challenges for Computational

Intelligence," edited by W. Duch and J. Mandziuk, is available online. Springer Publishing Company, 2007.

4. Fourth, U. Aickelin et al. ; PJ Bentley et al; S. Cayzer et al Is there a connection between AIS and IDS in Danger Theory? Pages 147–155, 2003, Second International Conference on Artificial Immune Systems.
5. "Evolutionary Algorithms in Theory and Practice" by T. Back. New York: Oxford University Press, 1996.